

## Invasion of non-native plant to the forest of the Cumberland Plateau and Mountain Region

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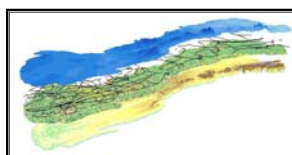
### INTRODUCTION:

#### Assessing Invasive Plants



- Invasion is not new phenomena; it is part of an evolutionary process.
- However, recently invasions have been greatly accelerated due to human influences.
- They are now considered a form of global change as they are occurring at an unprecedented rate across the globe.
- As our impact on the landscape changes the composition of 'natural' areas, it is important that we integrate technology to assist in active management.

### INTRODUCTION: GIS and Statistical Modeling



- Invasions are influenced by landscape pattern and scale.
- Tools that integrate space, time and scale are essential to understanding the underlying processes.
- Geographical Information Systems (GIS) is a tool that integrates these components and can be used to manage, analyze and disseminate spatial information.
- Relationships often not linear thus non-parametric modeling techniques are need.



#### Cumberland Plateau & Mountain Region



- 59,000 square kilometres
- One of the most diverse woody plant communities in the eastern United States.
- Forest resources are a major part of the economy.
- 70 % of the land in this area is forested, with over 75 % of this in hardwoods.

### RESEARCH QUESTIONS:

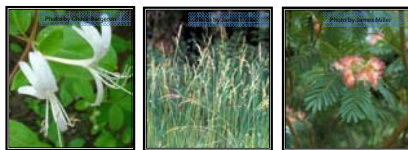
1. What is the probable distribution of three invasive species (Japanese honeysuckle, tall fescue and mimosa) in Cumberland Plateau and Mountain region ?
2. What is the relative importance of landscape drivers on the distribution of these invasive plants?
  - a) environment (e.g.. elevation, water sources etc.)
  - b) anthropogenic (e.g. distance to human features, management etc.)
3. How does plant occurrence affect our ability to model the probable distribution.

### DATA: Forest Inventory Analysis

- USDA Forest Service program: collects, analyses, and reports information on the status, trends and conditions of forests within the U.S.
- There is an extension of the Forest Inventory Analysis database that focuses on invasive plants.
- Invasives identified: 4 tree species of invasives, 7 shrubs, 7 vines, 5 grasses and 2 forbs, for a total of 25 invasive plants in the Cumberland Plateau and Mountain region.

Name	Occurrence
Tree of Heaven	39
Mimosa	45
Princess Tree	8
Russian olive	1
Autumn Olive	9
Burning Bush	1
Chinese Privet	234
Japanese Privet	71
Bush honeysuckles	10
Sacred Bamboo	5
Nonnative Roses	141
Asian Bittersweet	1
Chinese Yarn	7
Winter Creeper	2
Japanese Honeysuckle	579
Kudzu	11
Periwinkles	1
Japanese Wisteria	2
Tall Fescue	93
Cogongrass	1
Nepalese Browntop	49
Chinese Silvergrass	10
Nonnative Bamboos	9
Shrubby Lespedeza	9
Chinese Lespedeza	25

### FOCAL SPECIES:



Japanese Honeysuckle

Tall Fescue

Mimosa

Three species were chosen to study, based on:

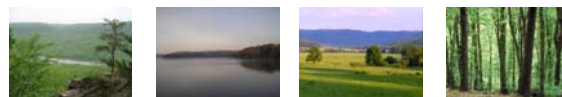
- Overall number of sites of occurrence
- Life forms

Species chosen were:

- Japanese honeysuckle (*Lonicera japonica*) [ $n = 579$ ][Vine]
- Tall fescue (*Lolium arundinaceum*) [ $n = 93$ ][Grass]
- Mimosa (*Albizia julibrissin*), [ $n = 45$ ][Tree]

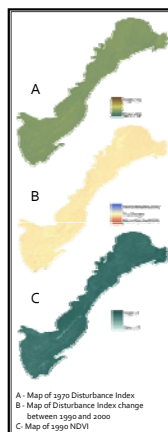
### DATA: Landscape Variables

- Landscape associated variables were derived from digital information.
- Landscape variables were categorized into six groups:
  - Landsat
  - Climate
  - Anthropogenic
  - Land use
  - Landform
  - Water
- Correlation within each group was assessed, and those with high correlation ( $>0.80$ ) were removed.



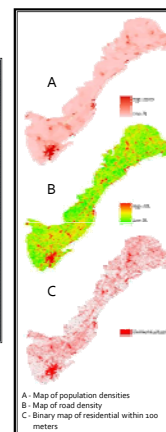
### DATA: Landsat

- Landsat imagery was used to assess forest disturbance.
- Two indices were calculated:
  - Normalized Difference Vegetation Index (NDVI)
  - Disturbance Index (DI)
- This was done for three time periods: 1970, 1990 and 2000.
- Landsat variables:
  - DI70
  - DI90
  - DI00
  - NDVI90
  - NDVI00



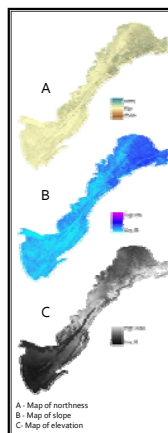
### DATA: Anthropogenic

- Invasive plants are often introduced and spread by people.
- Variables that represent human use were derived from road, census and land use data.
- Anthropogenic variables
  - Population
  - Road density
  - Road distance
  - Main road distance
  - Amount of developed area within 100 and 500 m



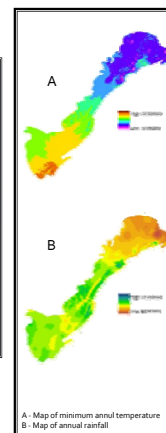
### DATA: Landform

- The landform variables were selected based their biological significance and correlation in other studies to plant distribution.
- 30m digital elevation model
- Environmental variables:
  - Northness
  - Eastness
  - Slope
  - Solar radiation
  - Curvature
  - Elevation



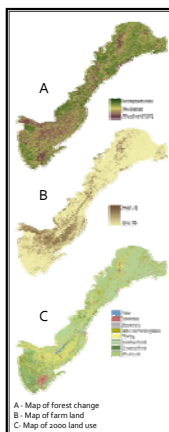
### DATA: Climate

- Environmental limitations of distribution are often highly influenced by climate, particularly rainfall and temperature.
- Monthly and annual temperature and rainfall maps were downloaded from PRISM.
- Highly correlated
- Climate variables:
  - Minimum annual temperature
  - Average annual rainfall



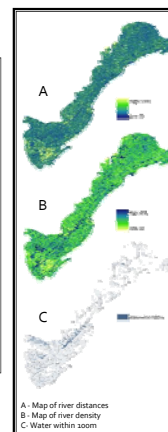
## DATA: Land Use

- Land use variables were extracted from the USGS classified the national land use cover datasets (NLCLU) for 2000 and 1990.
- Reclassified to 8 land uses based on Anderson's groupings.
- Land use variables:
  - Forest change (1990-2000) within a 500m buffer
  - Forest cover in 2000 within 100m buffer
  - Farming in 2000 within 500m buffer
  - Categorical variables of land use in 1990
  - Categorical variables of land use in 2000



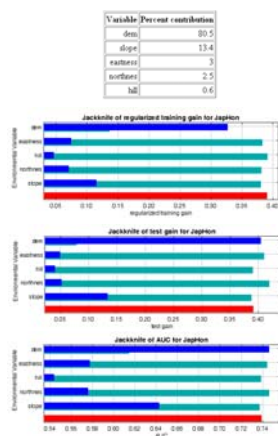
## DATA: Water

- Streams may affect the distribution and establishment of plants by influencing seed dispersal and moisture availability.
- National river shape files and water bodies defined in the NLCLU database.
- Water variables:
  - River distance
  - River density
  - Amount of water within 100m buffer
  - Amount of water within 500m buffer



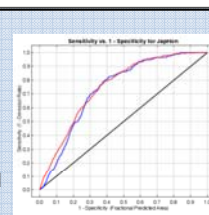
## MAXENT MODELS:

- Each group of variables (i.e. Landsat) was modeled using backward selection techniques.
- Full model was developed from all selected for each group, with only the significant variables kept in the final model.



## MAXENT MODELS:

- The omission rate, Cohen's Kappa and AUC were used to assess the reliability and validity of models.
- Evaluation statistics were calculated for both training (70%) and withheld (30%) data.
- Binary occurrence maps were developed using a threshold value that maximized the sum of sensitivity and specificity.



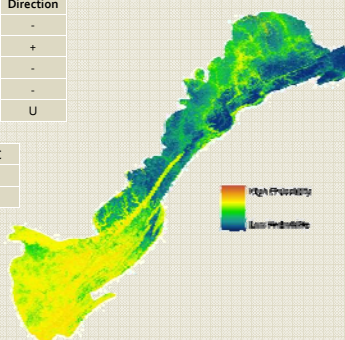
Variable threshold	Optimal threshold	Description	Threshold predicted area	Training omission rate	Test omission rate	P-value
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0.000	0.000	Random variable value 3	0.000	0.000	0.000	1.0000
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0.000	0.000	Random variable value 45	0.000	0.000	0.000	1.0000
0.000	0.000	Random variable value 46	0.000	0.000	0.000	1.0000
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0.000	0.000	Random variable value 62	0.000	0.000	0.000	1.0000
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0.000	0.000	Random variable value 66	0.000	0.000	0.000	1.0000
0.000	0.000	Random variable value 67	0.000	0.000	0.000	1.0000
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0.000	0.000	Random variable value 69	0.000	0.000	0.000	1.0000
0.000	0.000	Random variable value 70	0.000	0.000	0.000	1.0000
0.000	0.000	Random variable value 71	0.000	0.000	0.000	1.0000
0.000	0.000	Random variable value 72	0.000	0.000	0.000	1.0000
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0.000	0.000	Random variable value 99	0.000	0.000	0.000	1.0000
0.000	0.000	Random variable value 100	0.000	0.000	0.000	1.0000

## RESULTS : Japanese Honey Suckle

Variables	%	Direction
Elevation	58	-
Minimum Temperature	22	+
Amount of forest within 500m	10	-
Slope	6	-
Distance to Main Road	4	U

	Omission	Kappa	AUC
Training	0.20	0.50	0.87
Test	0.18	0.55	0.87

Cut off	0.42
Proportion of Occurrence	0.50

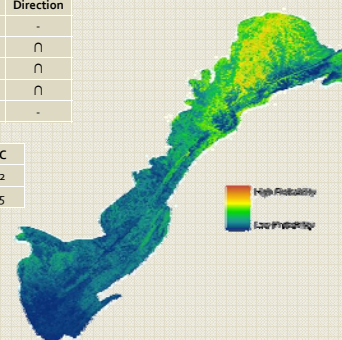


## RESULTS : Tall Fescue

Variables	%	Direction
Minimum Temperature	54	-
Elevation	19	∩
Farming within 500m	10	∩
Annual Rainfall	10	∩
Northness	7	-

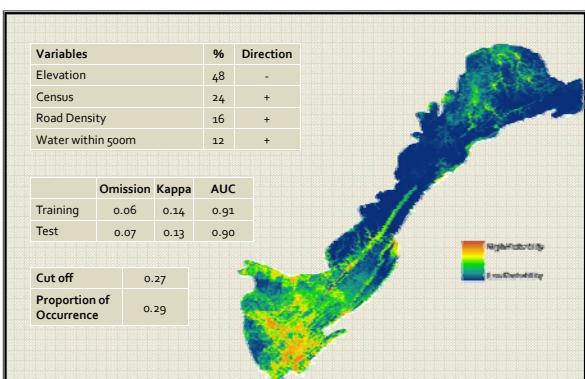
	Omission	Kappa	AUC
Training	0.25	0.19	0.82
Test	0.36	0.14	0.75

Cut off	0.42
Proportion of Occurrence	0.16





## RESULTS : Mimosa

RESULTS :  
Variable comparison

- Environmental variables dominated the models.
- Elevation in was used for three species.
- The single dominant variable either elevation or minimum temperature, both environmental variable.
- Some anthropogenic effects in all models.

	Japanese honeysuckle		Tall Fescue		Mimosa	
Variables	%	Direction	%	Direction	%	Direction
Elevation	58	-	19	∩	48	-
Slope	6	-				
Northness			7	-		
Annual Rainfall			10	∩		
Minimum Temperature	22	+	54	-		
Water within 500m					12	+
Total	86		90		60	

	Japanese honeysuckle		Tall Fescue		Mimosa	
Variables	%	Direction	%	Direction	%	Direction
Forest within 500m	10	-				
Farming within 500m			10	∩		
Distance to Main Road	4	U				
Road Density					16	+
Census					24	+
Total	14		10		40	

## RESULTS : Density comparison

- All models showed increasing spread of the species
- Omission rates were good, thus prediction of occurrence is good
- Kappa was good for Japanese honeysuckle but poor for tall fescue and mimosa, this takes into account absences, suggests these models don't predict absences well.
- AUC was reasonable for all models

	Proportion of sites with occurrence	Predicted Area	Omission	Kappa	AUC
Japanese honeysuckle	30%	50%	0.20	0.50	0.87
Tall Fescue	5%	16%	0.25	0.19	0.82
Mimosa	2%	29%	0.06	0.14	0.91

## DISCUSSION:

- More about habitat requirements (environmental) than anthropogenic activities

## ACKNOWLEDGEMENTS:

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Questions?

